



Supercomputer RAS via Informatics

(RAS – Reliability, Availability, Serviceability)

**CIS External Review
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Jon Stearley
jrstear@sandia.gov

**Sandia National Laboratories
Dept. 9224**



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Plenty of Faults, Plenty of Logs, But Actionable Information is Elusive

- **Event logs are a ubiquitous source of system feedback, but are notoriously free in format**
- **Supercomputers have many points of failure, with complex and dynamic interdependencies**
- ∴ **Identifying the root cause of faults in supercomputers is difficult (and expensive)**

Leverage recent advances in informatics!



Impact

- Inspection of system logs is fundamental to debugging – increased capability to quickly extract meaningful information **WILL** impact MTTR (mean time to repair) and **MAY** impact MTBF (mean time between failure). Red Storm is principal impact target.
- An analysis system which accepts any time-stamped sequence of free-text messages will:
 - NOT be device specific: Computer, Network Switch; Linux, TOS, Catamount; Cplant, ASCI Red, **Red Storm**
 - NOT be application specific: RAS, security, others...



Charter Statement

With the specific goal of increasing supercomputer RAS, we intend to produce a machine-learning analysis system which enables content-novice analysts to efficiently understand evolving trends, identify anomalies, and investigate cause-effect hypotheses in large multiple-source log sets.

Natural Language
Analysts

Supercomputer
Users



Computer Security
Analysts

Systems
Administrators



Automated Message Typing: Learning from Teiresias

What message content and occurrence rate is normal?

Teiresias (Bioinformatics code from IBM TJ Watson)

Two stages of operation:

1. **Scanning:** enumerate all elementary patterns of at least L/W specificity (I.e. find all phrases of length W words, where at least L of them never change (the others do change))
2. **Convolution:** combine elementary patterns into maximal irredundant “motifs” (leverage property of “downward closure”)

Sandia action: convert logs in/out of teiresias-acceptable format, cluster resulting “message templates” by time statistics

Result: automatically-generated “message templates”, sorted into three content categories: common, deviant, anomalous



Interactive Review via Logview

#	label	k	period	stddev	motif
L115	32	1	3		HWRPB cycle frequency (462962962) seems inaccurate - using the measured value of * Hz
L75	26	1	1		rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L76	6	3	6		rte-init: Found a LANai type * with 2097152 bytes (2048kB) of memory unit 0
L57	13	3600	0		named: XSTATS * 1007338854 RR=* RNXD=* RFwdR=* RDupR=* RFail=* RFErr=0 RErr=* RAXFR=76 RLan
L0	44	0	0		NOCLASS

☐ View lines in original (ungrouped) order

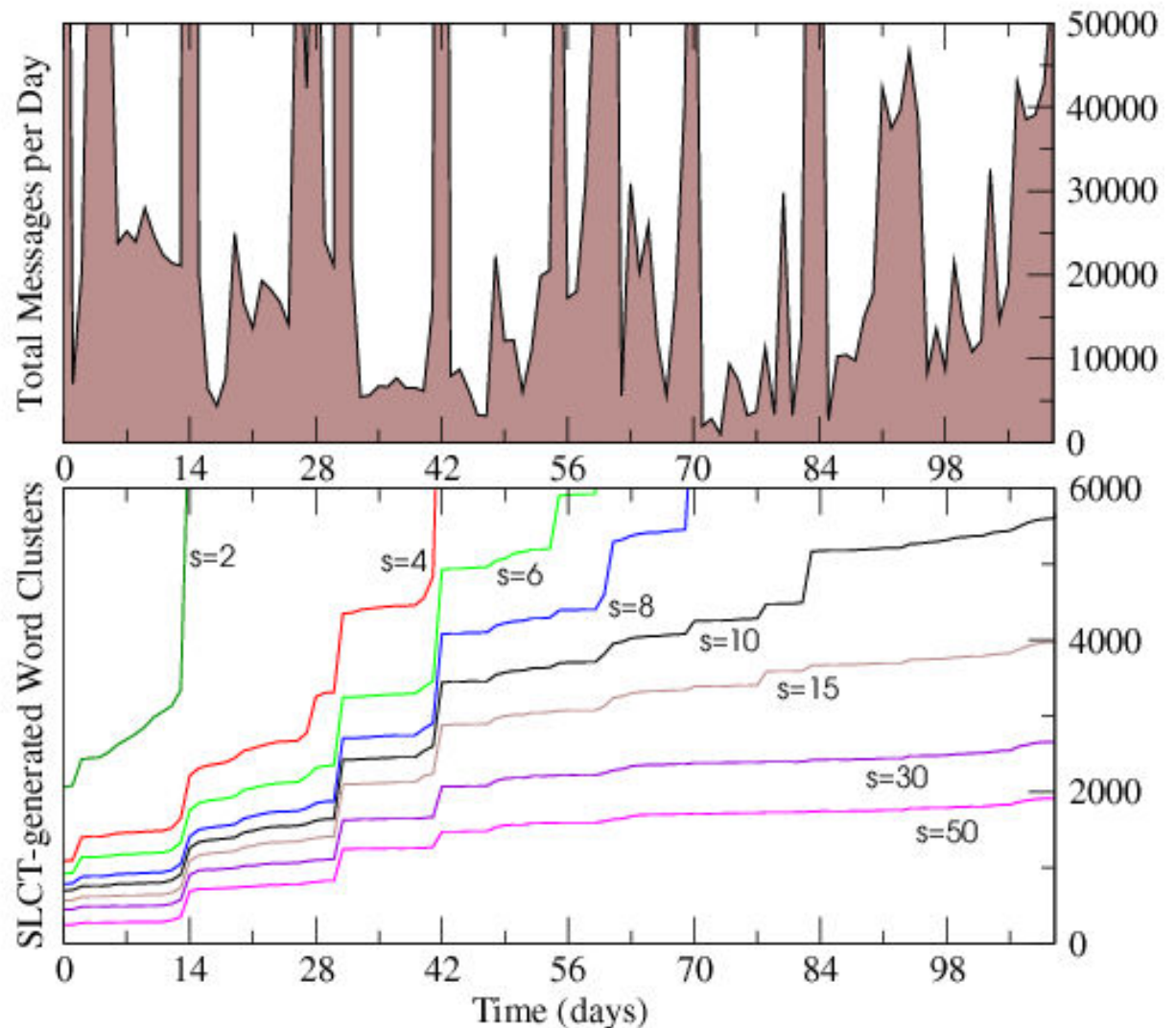
L75	Nov 25 17:53:25	src@node/if-0.n-3.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:25	src@node/if-0.n-23.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:25	src@node/if-0.n-4.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:27	src@node/if-0.n-11.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:29	src@node/if-0.n-2.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:29	src@node/if-0.n-17.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:30	src@node/if-0.n-9.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:30	src@node/if-0.n-22.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:31	src@node/if-0.n-5.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L75	Nov 25 17:53:31	src@node/if-0.n-8.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 7.2 with 2097152 bytes (2048kB) of memory unit 0
L76			
L76	Nov 25 17:53:04	src@node/if-0.n-15.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L76	Nov 25 17:53:07	src@node/if-0.n-20.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L76	Nov 25 17:53:23	src@node/if-0.n-28.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L76	Nov 25 17:53:24	src@node/if-0.n-32.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L76	Nov 25 17:53:25	src@node/if-0.n-25.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L76	Nov 25 17:53:31	src@node/if-0.n-29.t-37/if-1.n-0.t-37	rte-init: Found a LANai type 9.0 with 2097152 bytes (2048kB) of memory unit 0
L0			
L0	Nov 25 00:22:26	src@node/if-0.n-28.t-37/if-1.n-0.t-37	TSUNAMI machine check: vector=0x630 pc=0xfffffc000032f310 code=0x100000086
L0	Nov 25 00:22:26	src@node/if-0.n-28.t-37/if-1.n-0.t-37	machine check type: correctable ECC error (retryable)
L0	Nov 25 06:06:15	src@node/if-0.n-5.t-37/if-1.n-0.t-37	PCT-540[430]: ignoring ABORT_LOAD FIRST TRY from 797/2, unknown job ID 1733
L0	Nov 25 06:09:18	src@node/if-0.n-20.t-37/if-1.n-0.t-37	nfs: task 64052 can't get a request slot



Simple Logfile Clustering Tool (SLCT)

In contrast to Teiresias:

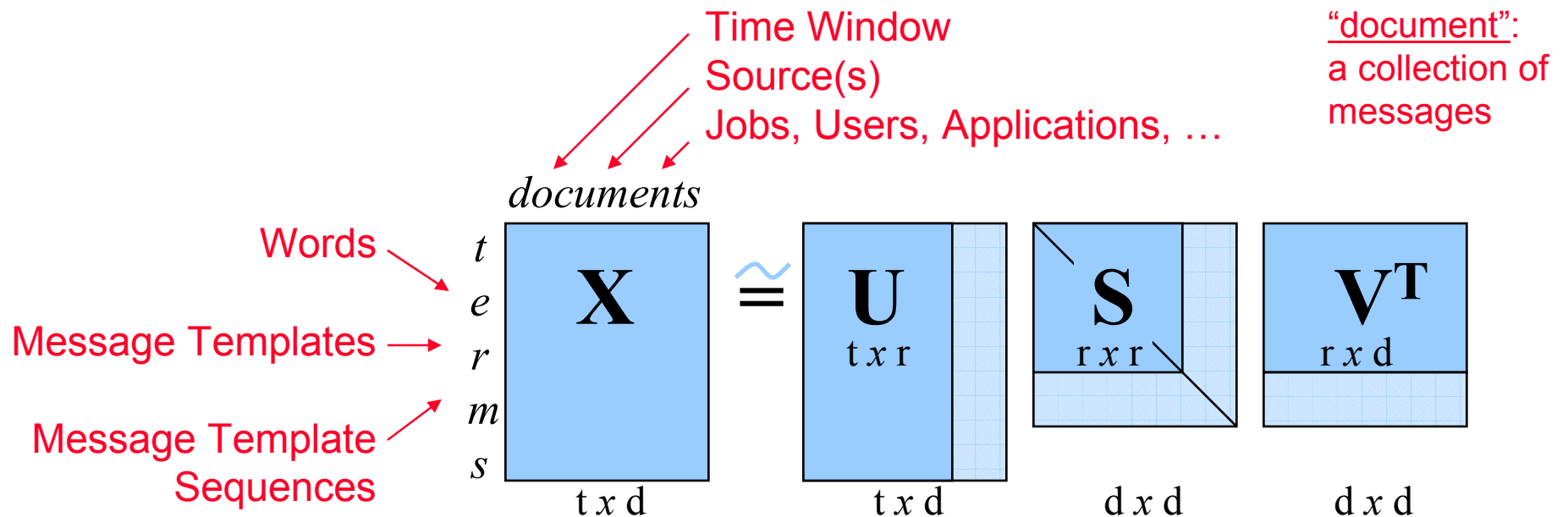
- Specifically designed to generate “message templates”
- Memory-efficient for log data
- Less effective anomaly categorization
- Open source ☺





Syslog Latent Semantic Analysis

Document similarity calculation using rank-reduced term space via Singular Value Decomposition.



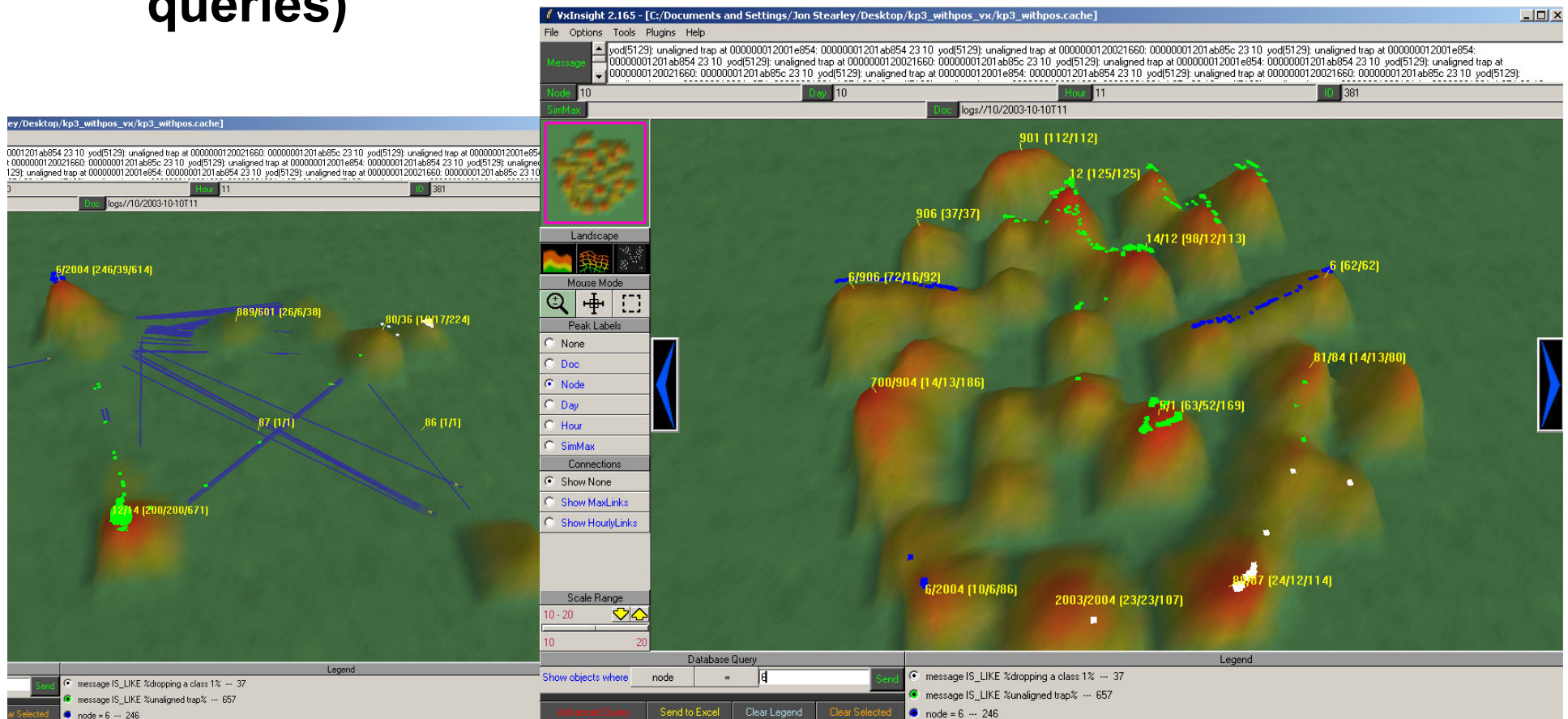
$$\text{Doc} \cdot \text{Doc similarity} = X^T X \approx X_r^T X_r \cong Y^T Y$$

$$\text{where } Y = \text{column normalize} [S_r V_r^T]$$



Exploration via VxInsight

- Sandia (9212) application, uses include patent and gene research (and now, syslogs)
- SQL database underneath (provides flexible queries)





Future Work

- **Study novelty rate**
 - Time rate of new message templates?
 - Rate of change of VxInsight landscape?
(must reach “steady state” to be practical)
- **Optimize parameters for effectiveness**
 - “Term” and “Document” creation parameters?
 - Are traditional LSA weighting functions best for this application?
 - Degree of Rank Reduction?
- **Quantitative effectiveness measure for test data sets**
- **Transition to RedStorm logs (RedStorm is primary impact target)**
- **Improve user interface**
 - Expose “most distinguishing phrase” as VxInsight labels
 - Reduce manual effort required per parameter change; utilize Sandia Text Analysis Library “STANLEY”
(Travis Bauer, 15241)

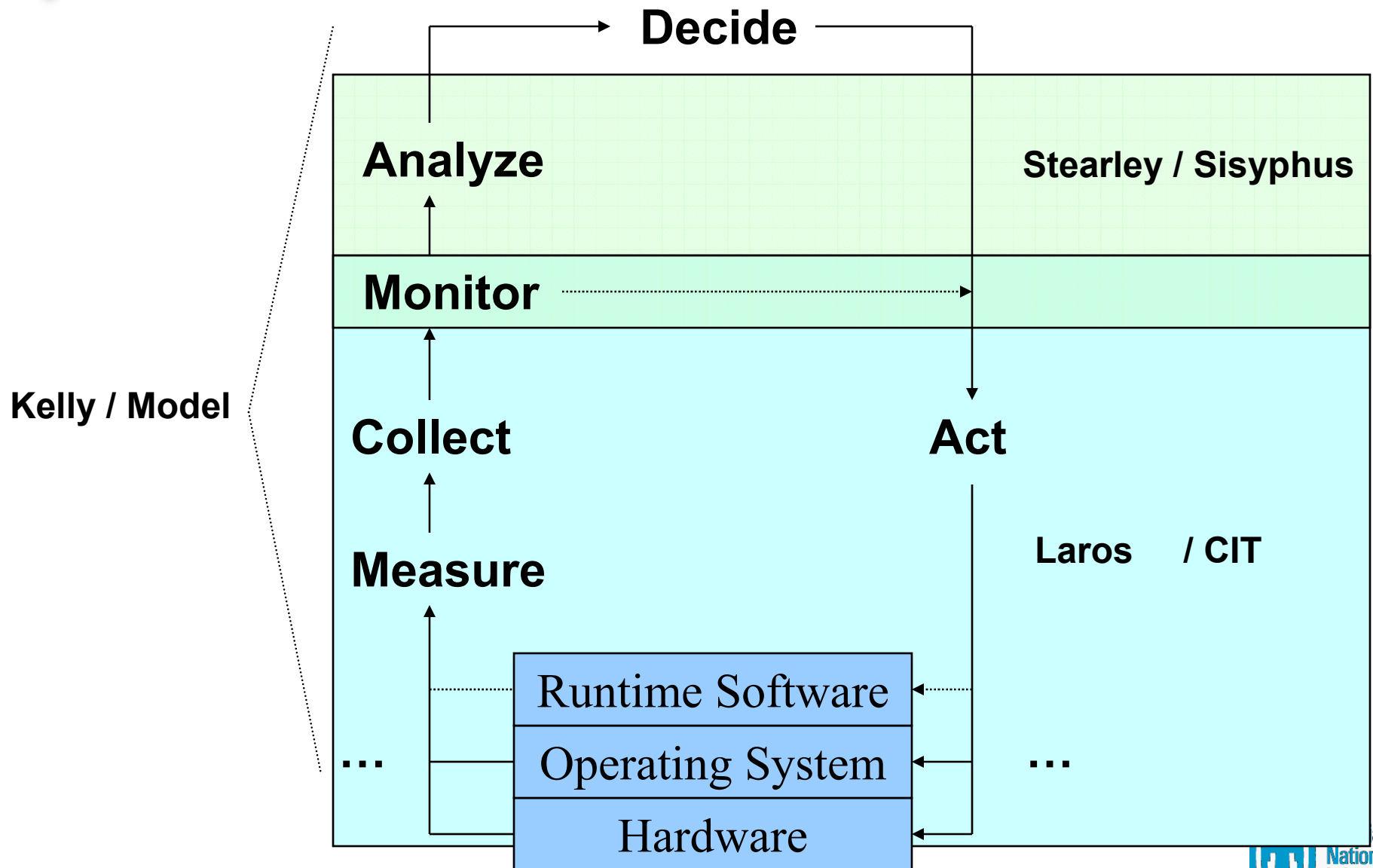


Backup Slides

Problem Statement

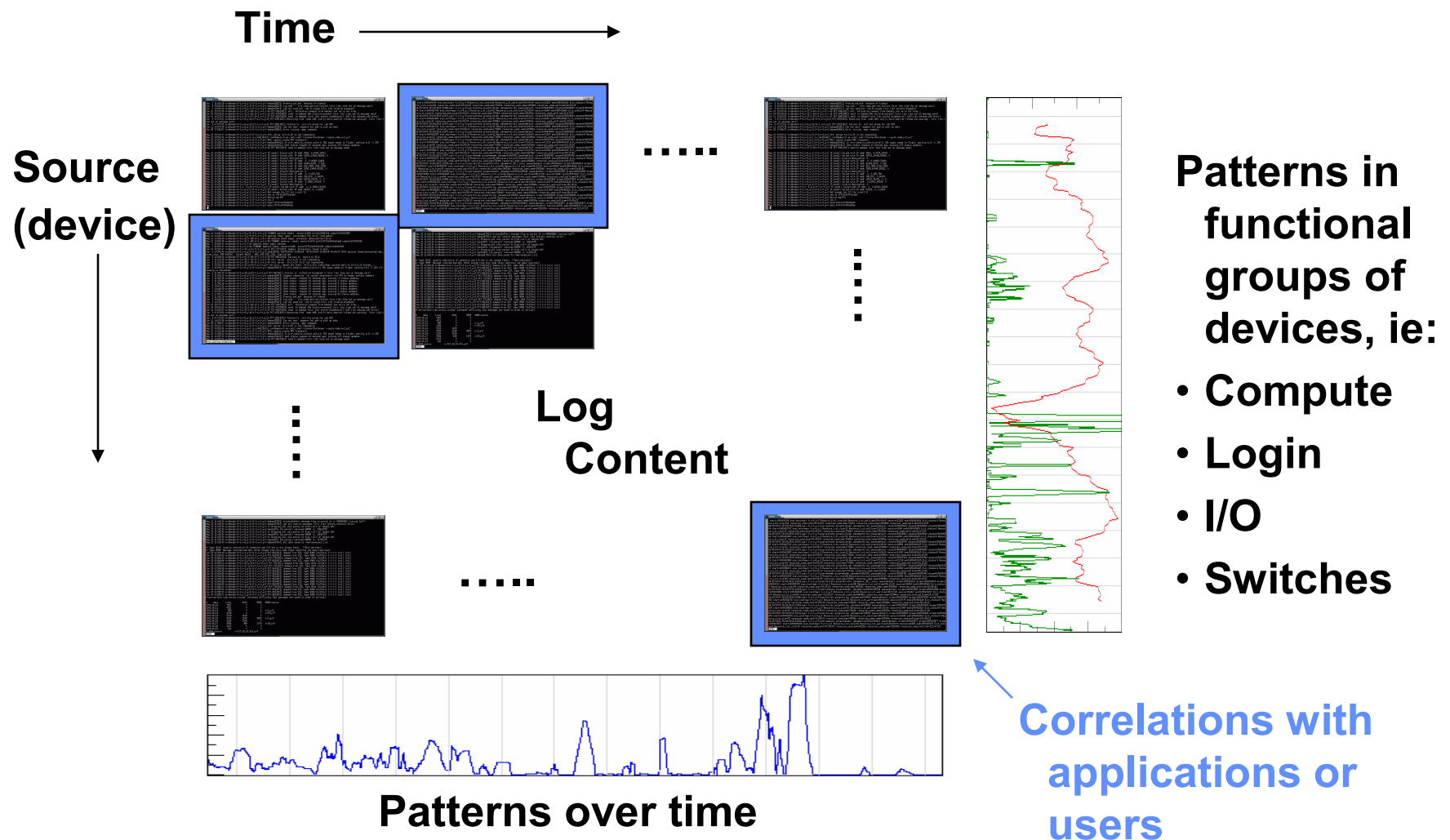


RAS Projects Context





Supercomputer RAS Via Informatics

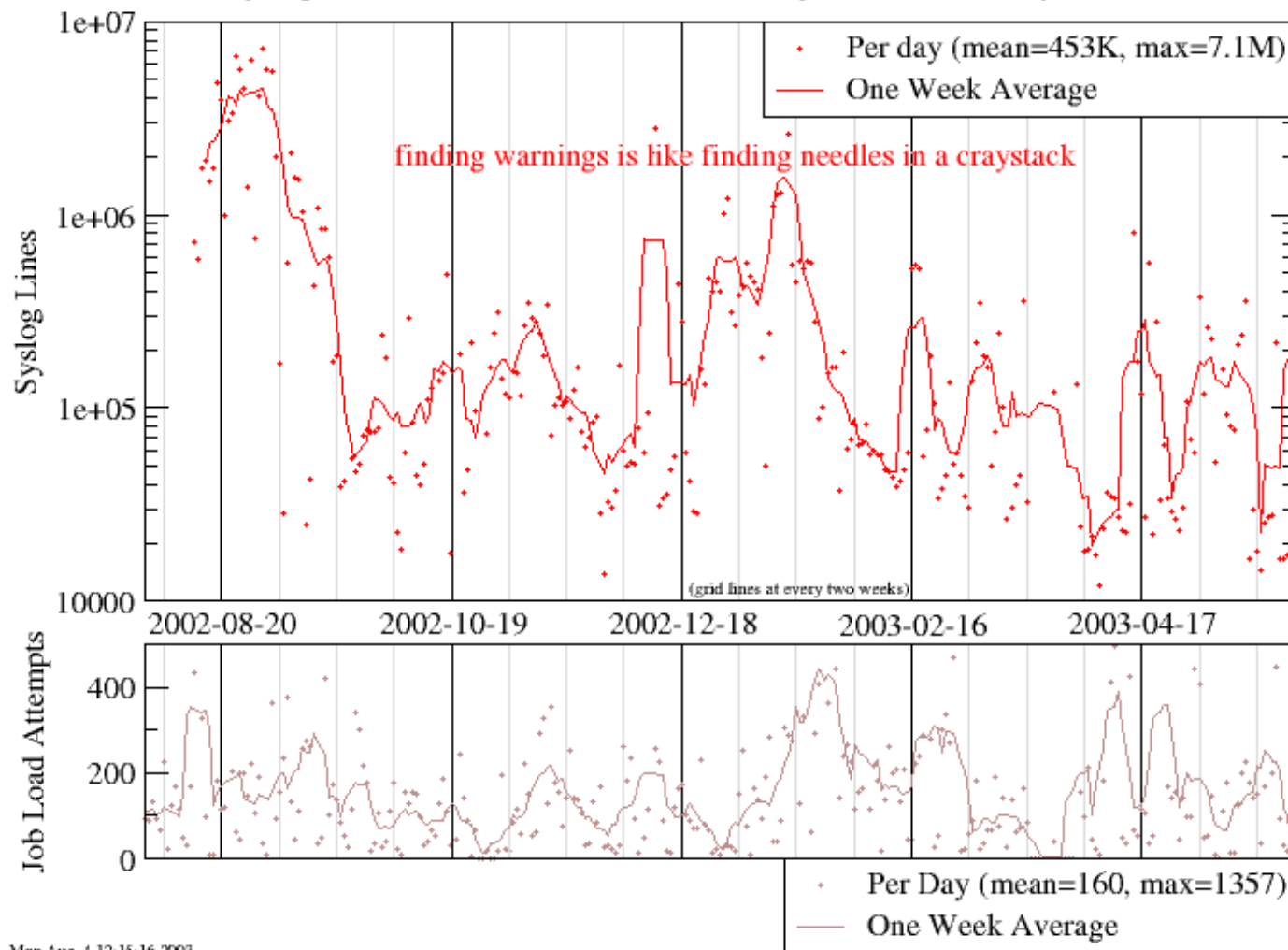




Plentiful Data, Elusive Information

Ross Syslog Lines

(roughly periodic with biweekly system reboots, slight correlation with job turnover rate)



Mon Aug 4 12:15:16 2003

Problem Statement



Research Context

- **Computer Security** - misuse and intrusion detection
- **Information Mgmt** - search engines, translation
- **Health** - gene and protein sequencing
- **National Security** - language modeling, antiterrorism

**Nobody is leveraging informatics
towards supercomputer RAS**
(upon which many of the above depend)



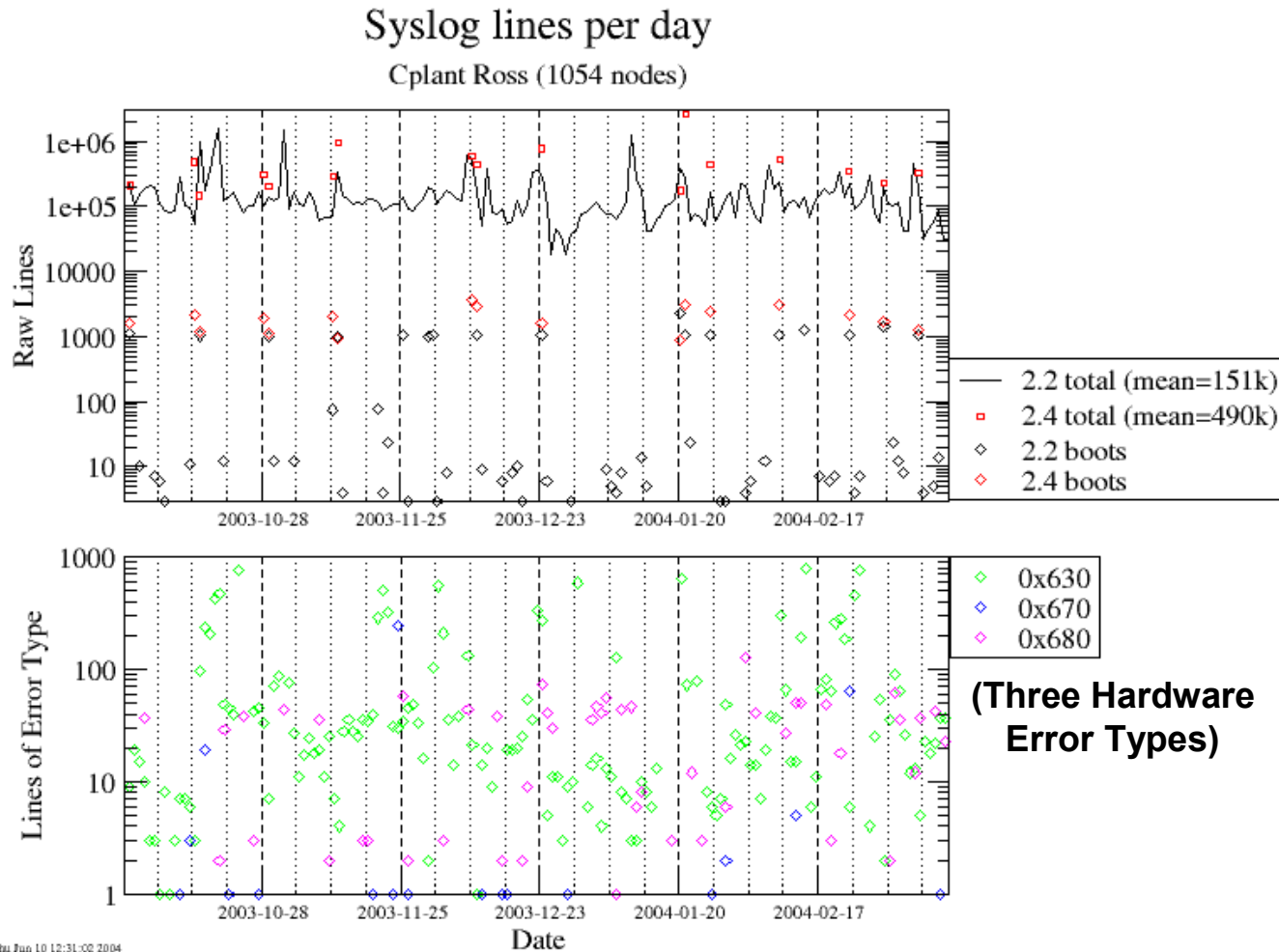
Analyst Thought Process

“What aspects of the message stream are strongly correlated with system malfunction or misuse?”

1. What message content and occurrence rate is normal?
2. What message groups are normal?
3. Can devices be classified by their output message stream?
4. Can users or applications be classified by their resulting message stream?
5. Are device-to-device and/or job-to-job log stream similarities sufficient to identify hardware or software failures?



Fault-Annotated Log Database



Thu Jan 10 12:31:02 2004

FY04 Progress

SQL Database
supports flexible
subset selection:

- By Device
- By Time
- By Job
- By User
- By Error or Message Type
- etc



Automated Grouping of Time-correlated Messages

What message groups are normal?

Approach:

Cluster message templates using their occurrence statistics:

1. Support
2. Inter-arrival median
3. Inter-arrival standard deviation

Results:

1. Detects periodic messages
2. Groups message groups which are rigidly time-correlated.



Automated Message Typing: Leveraging SLCT

SLCT – Simple Logfile Clustering Tool

Operates in three phases:

1. Frequent words –count all {position, word} tuples (pw), prune those occurring less than S times
2. Frequent messages – count all single-line pw “clusters” (“1w 2w 3_* 4_w”, ie message types)
3. Wildcard refinement (optional) – determine constant prefix or suffix for wildcards

Output and Categorization:

- pw clusters occurring at least S times (**common**)
- less specific pw clusters which, if joined with above pw clusters, occur at least S times (**common+deviant**)
- lines not matching either of the above (**rare**)



Term-Document Matrix

$$\begin{array}{c} \text{documents} \\ t \\ e \\ r \\ m \\ s \end{array} \begin{array}{c} \text{t} \times \text{d} \\ \mathbf{X} \end{array} = \text{column normalize} \left[\begin{array}{c} \text{Global} \\ \text{Term} \\ \text{Distribution} \\ \text{t} \times \text{t} \\ \mathbf{G} \end{array} \begin{array}{c} \text{Local} \\ \text{Term} \\ \text{Frequency} \\ \text{t} \times \text{d} \\ \mathbf{L} \end{array} \right]$$

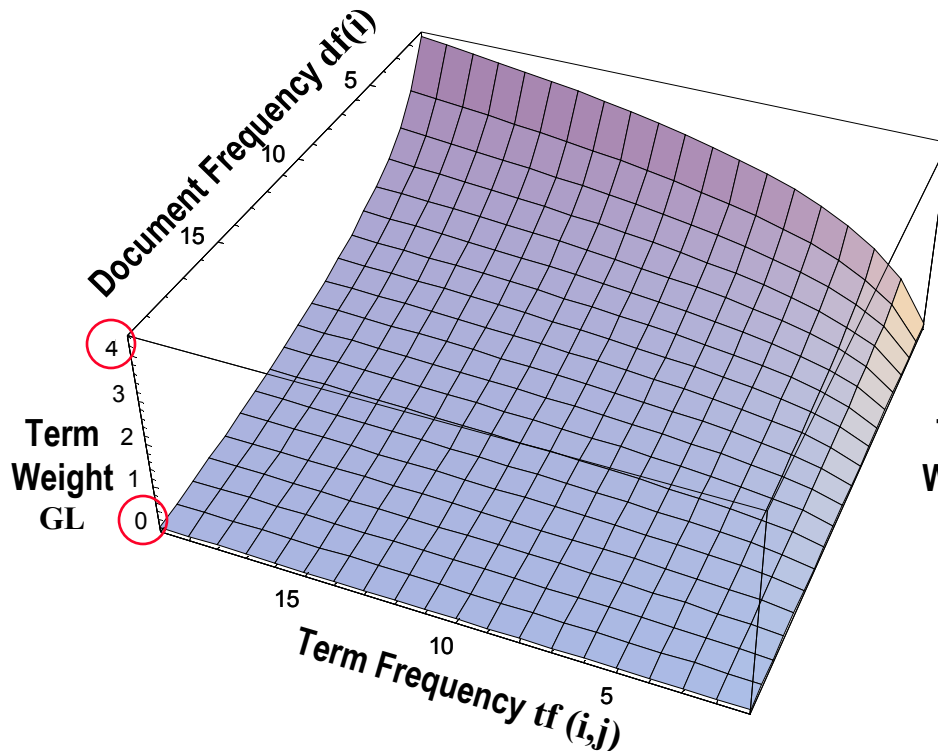
G and **L** are weighting functions.

Most simple case is $\mathbf{G}=\mathbf{I}$ and $\mathbf{L}=\mathbf{tf(i,j)}$,
where $\mathbf{tf(i,j)}$ is “term-frequency” of **i**’th term
in **j**’th document)



Term-Doc Matrix Weighting Functions

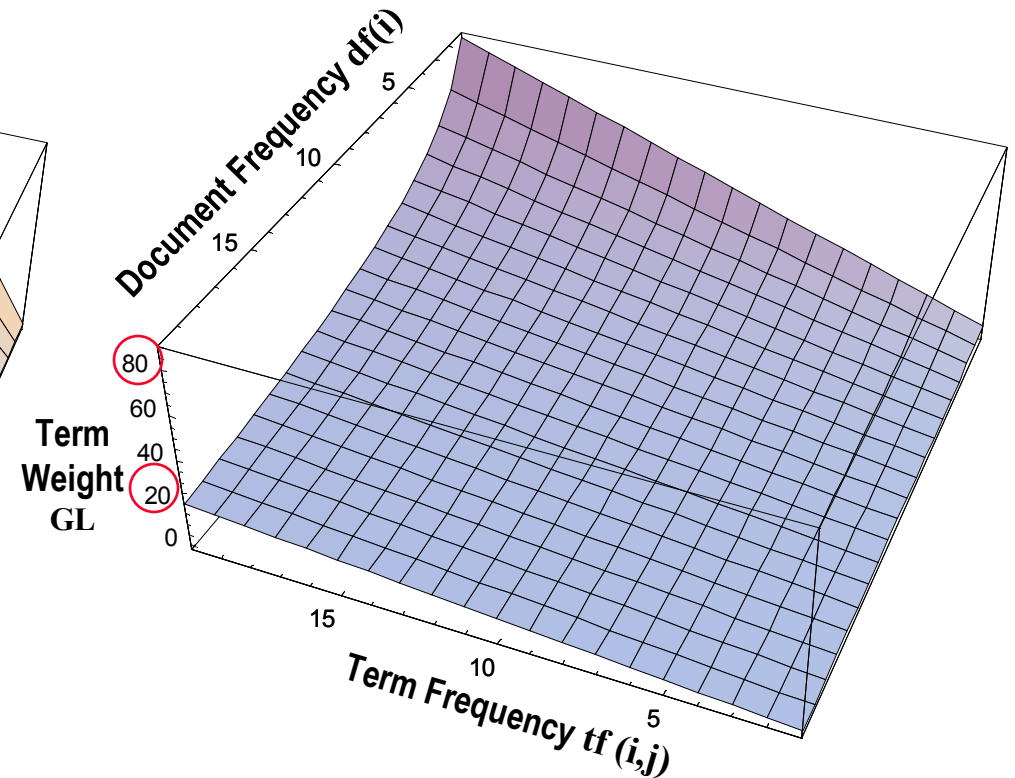
Log-Entropy



$$L(i, j) = \log(tf(i, j) + 1)$$

$$G(i) = 1 + \frac{1}{\log(n)} \sum_{j=1}^n p_{ij} \log(p_{ij}), \quad p_{ij} = \frac{tf(i, j)}{\sum_{j=1}^n tf(i, j)}$$

Inverse Document Frequency



$$L(i, j) = tf(i, j)$$

$$G(i) = \log\left(\frac{n}{df(i)} + 1\right)$$